## Intrinsic complexity in arithmetic (and algebra)

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JAF32, Athens, June 26, 2013

$$(\varepsilon) \quad \gcd(a, b) = \mathsf{if} \; (\mathsf{rem}(a, b) = 0) \; \mathsf{then} \; b \; \mathsf{else} \; \gcd(b, \mathsf{rem}(a, b))$$

where a = iq(a, b)b + rem(a, b)  $(0 \le rem(a, b) < b)$ 

 $\begin{aligned} \mathsf{calls}_{\{\mathsf{rem}\}}(\varepsilon, a, b) &= \mathsf{the number of divisions } \varepsilon \text{ needs to compute } \mathsf{gcd}(a, b) \\ &\leq 2\log(b) \qquad (a \geq b \geq 2) \end{aligned}$ 

▶ Is  $\varepsilon$  optimal for computing gcd(*a*, *b*) from {rem, =<sub>0</sub>}?

• 
$$a \perp b \iff \gcd(a, b) = 1$$

Is  $\varepsilon$  optimal for deciding coprimeness from {rem, =<sub>0</sub>, =<sub>1</sub>}?

• And is this true for all algorithms from  $\{\text{rem}, =_0, =_1\}$ ? **Conjecture**: For every algorithm  $\alpha$  which decides coprimeness from  $\{\text{rem}, =_0, =_1\}$ 

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)  $\operatorname{gcd}(a, b) = \operatorname{if} (\operatorname{rem}(a, b) = 0)$  then b else  $\operatorname{gcd}(b, \operatorname{rem}(a, b))$ 

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 $\mathsf{calls}_{\{\mathsf{rem}\}}(arepsilon, a, b) = \mathsf{the} \ \mathsf{number} \ \mathsf{of} \ \mathsf{divisions} \ arepsilon \ \ \mathsf{needs} \ \mathsf{to} \ \mathsf{compute} \ \ \mathsf{gcd}(a, b) \\ \leq 2 \log(b) \qquad (a \geq b \geq 2)$ 

▶ Is  $\varepsilon$  optimal for computing gcd(*a*, *b*) from {rem, =<sub>0</sub>}?

 $\bullet \ a \bot b \iff \gcd(a, b) = 1$ 

Is  $\varepsilon$  optimal for deciding coprimeness from  $\{\text{rem}, =_0, =_1\}$ ?

• And is this true for all algorithms from  $\{\text{rem}, =_0, =_1\}$ ? **Conjecture**: For every algorithm  $\alpha$  which decides coprimeness from  $\{\text{rem}, =_0, =_1\}$ 

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• A classical method for establishing lower bounds that restrict all algorithms assuming practically nothing about "what algorithms are":

**Horner's rule**: For any field F and  $n \ge 1$ , the value of a polynomial of degree n can be computed using no more than n multiplications and n additions in F:

$$a_0 + a_1 x + a_2 x^2 + \dots + a_n x^n = a_0 + x (a_1 + a_2 x + \dots + a_n x^{n-1})$$

#### **Theorem** (Pan 1966, (Winograd 1967, 1970))

Every algorithm from the complex field operations requires at least n multiplications/divisions and at least n additions/subtractions to compute  $a_0 + a_1x + a_2x^2 + \cdots + a_nx^n$  when  $\vec{a}, x$  are algebraically independent complex numbers (the generic case)

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#### Theorem (van den Dries)

If an algorithm lpha computes  $\gcd(x,y)$  from  $0,1,+,-,\mathsf{iq},\mathsf{rem},\cdot,<$  and

 $calls(\alpha, x, y) =$  the number of calls to the primitives  $\alpha$  makes to compute gcd(x, y),

then for all 
$$a > b$$
 such that  $a^2 = 2b^2 + 1$  (Pell pairs),  
calls $(\alpha, a + 1, b) \ge \frac{1}{4}\sqrt{\log \log b}$ 

... because it takes at least that many applications of the primitives to construct the value gcd(a + 1, b) when (a, b) is a Pell pair

 This method cannot yield lower bounds for decision problems (because their output (t or ff) is available with no computation)
 and it is open whether algorithms that decide coprimeness from these primitives (which include multiplication) must execute O(√log log max(x, y)) operations on an infinite set of inputs

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A (partial) structure is a tuple A = (A, Φ<sup>A</sup>) where Φ is a set of function and relation symbols and Φ<sup>A</sup> = {φ<sup>A</sup>}<sub>φ∈Φ</sub>, where with s<sub>φ</sub> ∈ {a, boole}, A<sub>a</sub> = A, A<sub>boole</sub> = {tt, ff},

$$\phi^{\mathbf{A}}: A^{n_{\phi}} \rightharpoonup A_{s_{\phi}}$$
 i.e.,  $\phi^{\mathbf{A}}: A^{n_{\phi}} \rightharpoonup A$  or  $\phi^{\mathbf{A}}: A^{n_{\phi}} \rightharpoonup \{\mathtt{t}, \mathtt{ff}\}$ 

- N = (N, 0, 1, +, ·, =), the standard structure of arithmetic
  N<sub>ε</sub> = (N, rem, =<sub>0</sub>, =<sub>1</sub>), the Euclidean structure
  N<sub>ε</sub> ↾ U = (U, rem ↾ U, =<sub>0</sub>↾ U, =<sub>1</sub>↾ U) where U ⊆ N and (f ↾ U)(x, y) = w ⇔ x ∈ U<sup>n</sup>, w ∈ U<sub>s</sub> & f(x) = w
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eqdiag( $\mathbf{A}$ ) = {( $\phi, \vec{x}, w$ ) :  $\vec{x} \in A^{n_{\phi}}, w \in A_{s_{\phi}}$ , and  $\phi^{\mathbf{A}}(\vec{x}) = w$ }

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With each structure  $\mathbf{A} = (A, \mathbf{\Phi})$ , each  $\Phi_0 \subseteq \Phi$  and each (partial) function or relation  $f : A^n \rightarrow A_s$  we will associate a partial function

$$\vec{x} \mapsto \mathsf{calls}_{\Phi_0}(\mathbf{A}, f, \vec{x}) \in \mathbb{N}$$
  $(f(\vec{x}) \downarrow)$ 

such that:

 $(\star)$  If  $\alpha$  is any algorithm from  $\mathbf{\Phi}$  which computes f, then

 $\operatorname{calls}_{\Phi_0}(\mathbf{A}, f, \vec{x}) \leq \operatorname{calls}_{\Phi_0}(\alpha, \vec{x}) \quad (f(\vec{x}) \downarrow)$ 

- (\*) is not trivial: in some important examples in arithmetic and algebra it yields the best known lower bound results
- (\*) is a theorem for concrete algorithms specified by the usual computation models; it is plausible for all algorithms from Φ
- The results are about several natural complexity measures on algorithms from primitives, not only "the number of calls to Φ<sub>0</sub>"
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With each structure  $\mathbf{A} = (A, \mathbf{\Phi})$ , each  $\Phi_0 \subseteq \Phi$  and each (partial) function or relation  $f : A^n \rightharpoonup A_s$  we will associate a partial function

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## Substructures and homomorphisms

#### Substructures (pieces):

$$\begin{split} \mathbf{U} &\subseteq_{\rho} \mathbf{A} = (A, \mathbf{\Phi}) \iff U \subseteq A \And \operatorname{eqdiag}(\mathbf{U}) \subseteq \operatorname{eqdiag}(\mathbf{A}) \\ \iff U \subseteq A \And (\forall \phi \in \mathbf{\Phi}) [\phi^{\mathbf{U}} \sqsubseteq \phi^{\mathbf{A}}] \end{split}$$

Substructures may be finite and not closed under  ${f \Phi}$ 

▶ A homomorphism  $\pi : \mathbf{U} \rightarrow \mathbf{V}$  is any  $\pi : U \rightarrow V$  such that for all  $\phi \in \Phi, x_1, \dots, x_n \in U, w \in U_s$ , (with  $\pi(\mathtt{t}) = \mathtt{t}, \pi(\mathtt{ff}) = \mathtt{ff}$ )

$$\phi^{\mathsf{U}}(x_1,\ldots,x_n)=w\implies\phi^{\mathsf{V}}(\pi x_1,\ldots,\pi x_n)=\pi w$$

• May have  $x \neq y, \pi(x) = \pi(y)$ , unless  $(=, x, y, \text{ff}) \in \mathsf{eqdiag}(\mathsf{U})$ 

•  $\pi$  is an embedding if it is injective (in which case it preserves  $\neq$ )

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An *n*-ary algorithm  $\alpha$  of  $\mathbf{A} = (A, \Phi)$  (or from  $\Phi$ ) "computes" some *n*-ary partial function or relation

$$\overline{\alpha} = \overline{\alpha}^{\mathbf{A}} : A^n \rightharpoonup A_s$$

#### using the primitives in $\boldsymbol{\Phi}$ as oracles and nothing else about $\boldsymbol{\mathsf{A}}$

We understand this to mean that in the course of a "computation" of  $\overline{\alpha}(\vec{x})$ , the algorithm may request from the oracle for any  $\phi^{A}$  any particular value  $\phi^{A}(\vec{u})$ , for arguments  $\vec{u}$  which it has already computed from  $\vec{x}$ , and that if the oracles cooperate, then "the computation" of  $\overline{\alpha}(\vec{x})$  is completed in a finite number of "steps"

- The notion of a uniform process attempts to capture minimally (in the style of abstract model theory) these aspects of algorithms from primitives
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Uniform processes: I The Locality Axiom

A uniform process  $\alpha$  of arity n and sort s of a structure  $\mathbf{A} = (A, \Phi^{\mathbf{A}})$  assigns to each substructure  $\mathbf{U} \subseteq_p \mathbf{A}$  an n-ary partial function

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$$\mathbf{U} \vdash \alpha(\vec{x}) = w \iff \overline{\alpha}^{\mathbf{U}}(\vec{x}) = w,$$
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If  $\alpha$  is an n-ary uniform process of  $\mathbf{A}$ ,  $\mathbf{U}, \mathbf{V} \subseteq_{p} \mathbf{A}$ , and  $\pi : \mathbf{U} \to \mathbf{V}$  is a homomorphism, then

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#### $\mathbf{A} \vdash \alpha(\vec{x}) \downarrow \implies (\exists \mathbf{U} \subseteq_{\rho} \mathbf{A}) [\mathbf{U} \vdash_{c} \alpha(\vec{x}) \downarrow]$

► I The Locality Axiom:

A uniform process  $\alpha$  of arity n and sort s of a structure  $\mathbf{A} = (A, \Phi^{\mathbf{A}})$  assigns to each substructure  $\mathbf{U} \subseteq_p \mathbf{A}$  an n-ary partial function

$$\overline{\alpha}^{\mathsf{U}}: U^{n} \rightharpoonup U_{s}$$

It computes the partial function or relation  $\overline{\alpha}^{\mathbf{A}} : A^n \rightharpoonup A_s$ 

$$\mathbf{U}\vdash\alpha(\vec{x})\downarrow\iff\alpha^{\mathbf{U}}(\vec{x})\downarrow$$

▶ II The Homomorphism Axiom: If  $\mathbf{U}, \mathbf{V} \subseteq_p \mathbf{A}$  and  $\pi : \mathbf{U} \to \mathbf{V}$  is a homomorphism, then

$$\overline{\alpha}^{\mathsf{U}}(\vec{x}) = w \implies \overline{\alpha}^{\mathsf{V}}(\pi \vec{x}) = \pi w$$

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Yiannis N. Moschovakis: Intrinsic complexity in arithmetic (and algebra)

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Suppose  $f : A^n \rightharpoonup A_s$ ,  $f(\vec{x}) \downarrow$ ,  $\mathbf{U} \subseteq_p \mathbf{A}$ .

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The intrinsic complexities of f in **A** 

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# Deriving lower bounds by constructing homomorphisms

• The following two facts are immediate from the definitions:

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f lpha is a uniform process which computes  $\mathsf{f}:\mathsf{A}^n o\mathsf{A}_s$  in  $\mathsf{A},$  then

 $C_{\mu}(\mathbf{A}, f, \vec{x}) \leq C_{\mu}(\alpha, \vec{x}) \qquad (f(\vec{x})\downarrow)$ 

Lemma (The homomorphism test) Suppose  $\mu$  is a substructure norm (e.g., calls $_{\Phi_0}$ , size, depth) on a  $\Phi$ -structure **A**,  $f : A^n \rightarrow A_s$ ,  $f(\vec{x}) \downarrow$ , and

for every finite  $\mathbf{U} \subseteq_{p} \mathbf{A}$  which is generated by  $\vec{x}$ ,  $(f(\vec{x}) \in U_{s} \& \mu(\mathbf{U}, \vec{x}) < m) \implies (\exists \pi : \mathbf{U} \to \mathbf{A})[f(\pi(\vec{x})) \neq \pi(f(\vec{x}))];$ 

then  $C_{\mu}(\mathbf{A}, f, \vec{x}) \geq m$ .
# Deriving lower bounds by constructing homomorphisms

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### Lemma

If  $\alpha$  is a uniform process which computes  $f : A^n \rightharpoonup A_s$  in **A**, then

$$C_{\mu}(\mathbf{A}, f, \vec{x}) \leq C_{\mu}(\alpha, \vec{x}) \qquad (f(\vec{x})\downarrow)$$

# Lemma (The homomorphism test)

Suppose  $\mu$  is a substructure norm (e.g., calls $_{\Phi_0}$ , size, depth) on a  $\Phi$ -structure **A**,  $f : A^n \to A_s$ ,  $f(\vec{x}) \downarrow$ , and

for every finite  $\mathbf{U} \subseteq_{p} \mathbf{A}$  which is generated by  $\vec{x}$ ,  $(f(\vec{x}) \in U_{s} \& \mu(\mathbf{U}, \vec{x}) < m) \implies (\exists \pi : \mathbf{U} \to \mathbf{A})[f(\pi(\vec{x})) \neq \pi(f(\vec{x}))];$ 

then  $C_{\mu}(\mathbf{A}, f, \vec{x}) \geq m$ .

Deriving lower bounds by constructing homomorphisms

• The following two facts are immediate from the definitions:

### Lemma

If  $\alpha$  is a uniform process which computes  $f : A^n \rightharpoonup A_s$  in **A**, then

$$C_{\mu}(\mathbf{A}, f, \vec{x}) \leq C_{\mu}(\alpha, \vec{x}) \qquad (f(\vec{x})\downarrow)$$

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# A lower bound for coprimeness on $\ensuremath{\mathbb{N}}$

 $\textbf{A}=(\mathbb{N},0,1,+,\dot{-},\mathsf{iq},\mathsf{rem},=,<,\Psi)\text{, }\Psi$  a finite set of Presburger functions

Theorem (van den Dries, ynm, 2004, 2009)

If  $\xi > 1$  is quadratic irrational, then for some r > 0 and all sufficiently large coprime (a, b),

$$\left|\xi - \frac{a}{b}\right| < \frac{1}{b^2} \implies \operatorname{depth}(\mathbf{A}, \mathbb{L}, a, b) \ge r \log \log a.$$
 (1)

In particular, the conclusion of (1) holds with some

- for positive Pell pairs (a, b) satisfying  $a^2 = 2b^2 + 1$  ( $\xi = \sqrt{2}$ )
- ▶ for Fibonacci pairs  $(F_{k+1}, F_k)$  with  $k \ge 3$   $(\xi = \frac{1}{2}(1 + \sqrt{5}))$

### Theorem (Pratt, unpublished)

There is a non-deterministic algorithm  $\varepsilon_{nd}$  of  $N_{\varepsilon}$  which decides coprimeness, is at least as effective as the Euclidean everywhere and

### $\operatorname{calls}(\varepsilon_{nd}, F_{k+1}, F_k) \leq K \log \log F_{k+1}$

The theorem is best possible from its hypotheses

Yiannis N. Moschovakis: Intrinsic complexity in arithmetic (and algebra)

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The theorem is best possible from its hypotheses

Given N, how good can a coprimeness algorithm be if we only insist that it works for and uses only N-bit numbers?

 $A = (\mathbb{N}, 0, 1, +, -, iq, rem, =, <, \Psi)$  as before. For any N, and any one of the intrinsic complexities as above, let

 $C_{\mu}(\mathbf{A}, f, 2^{N}) = \max\{C_{\mu}(\mathbf{A} \upharpoonright [0, 2^{N}), f, \vec{x}) : x_{1}, \dots, x_{n} < 2^{N}\}$ 

Theorem (van den Dries, ynm 2009) For some rational number r > 0 and all sufficiently large N,

calls $(\mathbf{A}, \bot, 2^N) \ge \operatorname{size}(\mathbf{A}, \bot, 2^N) \ge r \log N.$ 

### ▶ Non-uniform lower bound for depth( $\mathbf{A}, \bot, 2^N$ )?

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$$N_F(a_0,\ldots,a_n,x) \iff a_0+a_1x+a_2x^2+\cdots+a_nx^n=0$$

#### Theorem

Let F be the field of real or complex numbers. If  $n \ge 1$  and  $a_0, \ldots, a_n, x$  are algebraically independent in F, then: (1) calls<sub>{.,÷}</sub>(F, N<sub>F</sub>,  $\vec{a}, x$ ) = n

(2) calls<sub>{ $\cdot, \div, =$ </sub>}(*F*, *N<sub>F</sub>*, *ā*, *x*) = *n* + 1

- The method for constructing the required homomorphsms is an elaboration of Winograd's proof of the optimality of Horner's rule for poly evaluation
- It is quite different from the method used in arithmetic and requires a homomorphism which is not an embedding in (2)
- Due to Bürgisser and Lickteig (1992) for algebraic decision trees, along with much stronger results

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